

Two Strongly Truthful Mechanisms for Three Heterogeneous Agents Answering One Question

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Motivation Questions

How can we design mechanisms to collect agents' truthful reports without verification?

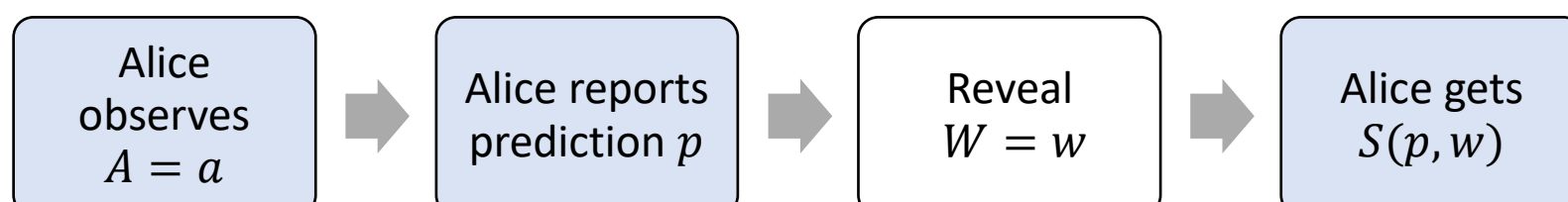
- three agents
- single item report
- asymmetric priors
- prior free

What is special about Bayesian Truth Serum?

- information score
- Log scoring rule

Proper scoring rule with Verification

Proper scoring rule $S(p, w) \in \mathbb{R}$



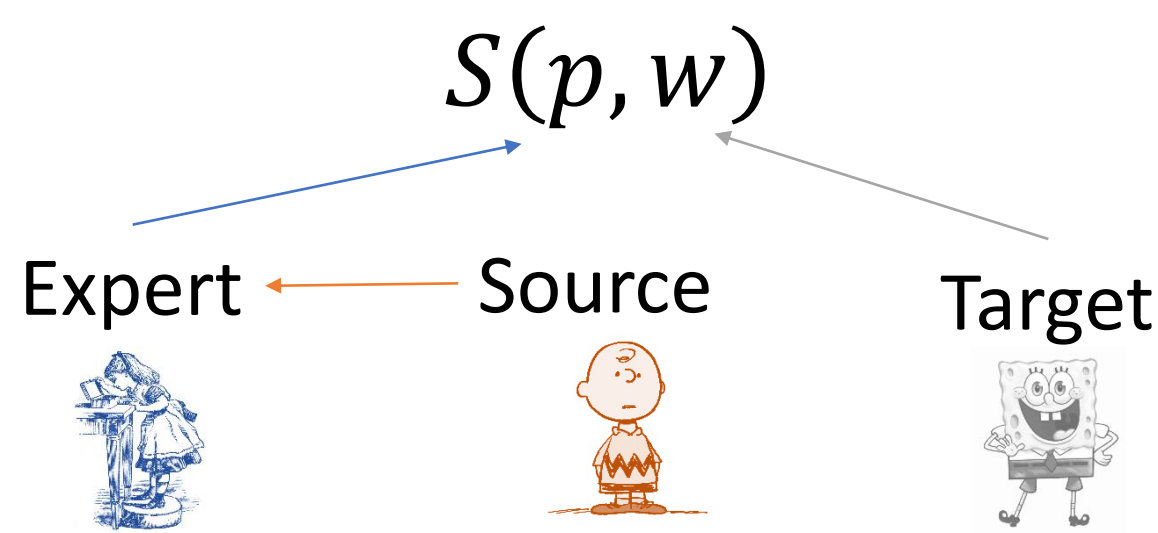
Log scoring rule $LS(p, w) = \log p(w)$

- **Proper:** truth-telling $P_{W|A}$ maximizes $LS(p, w)$
- Shannon mutual information: $\mathbb{E}_{W,A}[LS(P_{W|A}, W)] = MI(W; A)$
- **Chain rule:** If $p_i = P_{W|A_1 \dots A_i}$ for all i
 $\mathbb{E}[LS(p_i, W) - LS(p_{i-1}, W)] = MI(W; A_i | A_1, \dots, A_{i-1})$

Proper Scoring Rule without Verification

Agents can play one of three roles

- Target reports his signal and is predicted.
- Expert makes prediction the target's signal.
- Source provides information to the expert.



Differential Peer Prediction Mechanisms

Theorem. DPP Mechanism is strongly truthful:

1. The truth-telling strategy profile is a strict Bayesian Nash Equilibrium,
2. The ex-ante agent welfare in the truth-telling strategy profile is strictly better than all non-permutation strategy profiles,

if the common prior is second order stochastic relevant on a finite set with full support.

DPP mechanism

1. Bob and Charlie report their signals B and C
2. Set one as Target and the other as the Source randomly. Set Alice as the expert.
3. Alice reports her initial prediction Q on **Target's** signal.
4. Alice learns **Source's** signal.
5. Alice updates her improved prediction Q^+ on **Target's** signal.

Two ideas for payments

- Source incentive: Using Source's (Charlie's) report to improve prediction $Q \rightarrow Q^+$
- Target incentive: Using Target's (Bob's) report as the ground truth

Payment	Source DPP	Target DPP
Alice (Expert)	$LS(Q^+, B) + LS(Q, B)$	
Bob (Target)	0	$LS(Q^+, B) - LS(Q, B)$
Charlie (Source)	$LS(Q^+, B) - 3LS(Q, B)$	$-2LS(Q, B)$
Total	$2(LS(Q^+, B) - LS(Q, B)) = 2MI(C; B A)$	

Log scoring rule reversed

- Truthfulness the second argument?
- Reversed Log scoring rule
 $R(w) = LS(Q^+, w) - LS(Q, w)$
- Symmetry between Target and Source
 $\mathbb{E}[LS(Q^+, B) - LS(Q, B)] = MI(C; B | A)$

Connection to BTS and other mechanisms

Bayesian Truth Serum

- Each i agent reports her signal x_i and prediction p_i on other's signal.
- Prediction score: measure the quality of p_i
- Information score: $LS(Q^+, x_i) - LS(Q, x_i)$
Aggregated prediction One agent's prediction
- Target-incentive mechanism

Source-incentive mechanism

- Robust BTS,
- Shadowing mechanism,
- Knowledge-Free Peer Prediction